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## ABSTRACT

Rasch models for fundamental measurement in the psychological sciences are derived from the principle of specific objectivity, the requirement that the parameter value representing each component in a test situation be independent of the other components. The dichotomous Rasch model for two-faceted analysis, applicable to conventional paper-and-pencil tests, is constructed. The many-faceted Rasch model is also derived by means of a three-faceted example, comprising judges, examinees, test items, and a rating scale, which is applicable to many judging situations. Any other particular form of the many-faceted model may also be derived in a similar manner. (Author/TJH)

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## **Rasch Models from Objectivity:**

### **A Generalization**

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## Abstract:

Rasch models for fundamental measurement in the psychological sciences are derived from the principle of specific objectivity, the requirement that the parameter value representing each component in a test situation be independent of the other components. The dichotomous Rasch model for two-faceted analysis, applicable to conventional paper-and-pencil tests, is constructed. The many-faceted Rasch model is also derived by means of a three-faceted example, comprising judges, examinees, test items and a rating scale, which is applicable to many judging situations.

Key-words: Rasch measurement, Objectivity, Rating scales

## I. Introduction.

Georg Rasch has some wise words to say on the subject of objectivity which will guide us in this discussion:

"The concept of 'objectivity' raises fundamental problems in all sciences. For a statement to be scientific, 'objectivity' is required. However, exactly what 'objectivity' means is disputed among philosophers and I am not going to enter into that debate" (Rasch 1964 p.1).

"The comparison of any two subjects may be carried out in such a way that no other parameters are involved than those of the two subjects - neither the parameter of any other subject nor of any of the stimulus parameters.

"Similarly, any two stimuli may be compared independently of all other parameters than just those of the two stimuli - the parameters of all other stimuli as well as the parameters of the subjects having been replaced by observable numbers.

"It is suggested that comparisons carried out under such circumstances will be designated as 'specifically objective'. And the same term would seem appropriate for statements about the model structure which are independent of all of the parameters specified in the model, the unknown values of them being, in fact, irrelevant for the structure of the model" (Rasch 1966 p.21).

"A model is not meant to be true. Even in classical physics models are temporary - good enough for some purposes" (Rasch 1964 p. 2).

Our aim, then, is to characterize each component of a complex test situation by one parameter of fixed, but unknown, value which is objective, that is to say independent of the other parameters. These parameters are to be estimated, or measured, by means of a model which capitalizes on this invariance in the parameters.

## II. Objectivity for a two-faceted test.

Rasch developed the idea of objectivity in his analysis of an intelligence test taken by 1094 recruits to the Danish Army. (Rasch 1980 p.62). This was a conventional two-faceted test situation: the observations were the dichotomous (right-wrong, 1-0) results of interactions between the objects (examinees) and the agents (test items).

The derivation of the corresponding Rasch model, from objectivity, is included here as a demonstration of the principles involved following Wright (passim, in particular Wright and Linacre, 1987).

Consider a test in which two objects,  $O_m$  and  $O_n$ , respond to numerous replications of a dichotomously scored agent,  $A_i$ . The outcome of this test is depicted in Figure 1.

		Object $O_n$	
		0	1
Object $O_m$	0	F00	F10
	1	F01	F11

Figure 1. Outcome of the hypothetical administration of numerous replications of an agent to two objects. F10 is the count of the number of times that object  $O_n$  succeeded on the agent at the same time that object  $O_m$  failed.

The only way to determine the difference between two objects is to observe when they perform differently. In Figure 1, they have performed in the same way with counts of F00 and F11; they have performed differently with counts of F10 and F01. The comparison of  $O_n$  and  $O_m$  is thus observed through the comparison of F10 and F01, for which F00 and F11 provide a quantitative context. In whatever way the comparison of F10 and F01 is made, it must be independent of the length of the test. Thus, if the test were to be twice as long, with identically proportionate results, then the comparison between  $O_m$  and  $O_n$  would become a comparison between  $2 \cdot F10$  and  $2 \cdot F01$ . In order for this comparison to be independent of the arbitrary number of replications, the comparison between  $O_m$  and  $O_n$  can be estimated by

$$\frac{\text{Performance of } O_n}{\text{Performance of } O_m} \approx \frac{F10}{F01} \quad (1)$$

This ratio of frequencies is the empirical but stochastic manifestation of the ratio of the probabilities of the corresponding events. For objectivity, the responses made by the objects must not be influenced by each other and so must be independent. The probabilities corresponding to the frequencies in Figure 1 are specified in Figure 2. It is the counts in all 4 cells of

Figure 1 which enable the probabilities in Figure 2 to be estimated. Thus the performance of  $O_n$  and  $O_m$  can be defined in terms of the probabilities corresponding to  $P_{i0}$  and  $P_{i1}$ , giving

$$\frac{\text{Performance of } O_n}{\text{Performance of } O_m} = \frac{P_{n1} * P_{m0}}{P_{n0} * P_{m1}} \quad (2)$$

Object $O_n$	
0                      1	
Object $O_m$ 0	$P_{n0} * P_{m0}$ $P_{n1} * P_{m0}$
1	$P_{n0} * P_{m1}$ $P_{n1} * P_{m1}$

Figure 2. The probabilities of the outcome of the administration of agent  $A_i$  to objects  $O_n$  and  $O_m$ .  $P_{n1}$  is the probability of object  $O_n$  responding to agent  $A_i$  successfully.

By specific objectivity, the comparison of  $O_n$  and  $O_m$  must be independent of which particular agent is used in making the comparison. Therefore the comparison of their performance must be the same for agent  $A_i$  and for agent  $A_j$ . Therefore

$$\frac{\text{Performance of } O_n}{\text{Performance of } O_m} = \frac{P_{n1} * P_{m0}}{P_{n0} * P_{m1}} = \frac{P_{nj1} * P_{mj0}}{P_{nj0} * P_{mj1}} \quad (3)$$

Rearranging the terms,

$$\frac{P_{n1}}{P_{n0}} = \frac{P_{nj1}}{P_{nj0}} * \frac{P_{m1}}{P_{m0}} * \frac{P_{mj0}}{P_{mj1}} \quad (4)$$

This is true for all  $m, n, i, j$ . Thus it is also true if object  $O_m$  is object  $O_0$ , whose measure is defined to be at the local origin of the object measurement scale, and also true for agent  $A_0$ , whose calibration is defined to be at the local origin of the agent measurement scale. Then

$$\frac{P_{n1}}{P_{n0}} = \frac{P_{n01}}{P_{n00}} * \frac{P_{0i1}}{P_{0i0}} * \frac{P_{000}}{P_{001}} \quad (5)$$

But,  $P_{000}/P_{001}$  is a constant term dependent on the relative placement of the local origins of the object and agent sub-scales, so that, if they are defined to coincide, then  $P_{000} = P_{001} = 0.5$  and  $P_{000}/P_{001} = 1$ .

Then, once the local origin is defined,  $P_{n01}/P_{n00}$  is a ratio dependent only on object  $O_n$ , so that  $\log(P_{n01}/P_{n00})$  can be expressed as  $B_n$ . Similarly,  $\log(P_{0i0}/P_{0i1})$  is a ratio dependent only on agent  $A_i$ , which can be expressed as  $D_i$ . Equation (5) thus becomes, on taking logarithms and reparameterizing  $P_{ni1}$  to be  $P_{ni}$ , so that  $P_{ni0}$  is  $1 - P_{ni}$ ,

$$\log(P_{ni}/(1-P_{ni})) = B_n - D_i \quad (6)$$

which is the Rasch model for dichotomous, two-faceted data.

### III. Generalizing to many-faceted models using a three-faceted example.

The derivation of a particular form of the many-faceted model demonstrates the general principles by which any other particular form of the many-faceted model can also be derived. The particular model to be derived here is applicable to a three-faceted test in which each judge of a panel of judges awards a rating to each examinee on each item.

Consider the performance of two examinees,  $O_n$  and  $O_m$ , as rated by a particular judge,  $J_j$ , on replications of the same item,  $A_i$ . In whatever way the ratings were originally recorded, they have been recoded into  $K+1$  categories ordinally numbered from 0 to  $K$ , with each higher numbered category representing a higher level of perceived performance, and with each category having a non zero probability of occurrence.

Categories	Examinee	
	$k$	$l$
Examinee $O_m$	$F_{kk}$	$F_{lk}$
$l$	$F_{kl}$	$F_{ll}$

Figure 3.  $F_{kl}$  represents the count of the number of times that examinee  $O_n$  is awarded rating  $k$  and examinee  $O_m$  is rated a  $l$  by judge  $J_j$  across replications of item  $A_i$ , where  $k > l$ .

The administration of the numerous replications of item  $A_i$  is the "test". The performance levels of examinees  $O_n$  and  $O_m$  can be compared by their relative frequencies of being rated in the various categories of the rating scale. Following the procedure in the discussion of objectivity, let us summarize part of their performance by a 2x2 cross-tabulation of counts of ratings in categories  $k$  and  $l$  of the rating scale, chosen so that category  $k$  is numerically greater than category  $l$  and so represents a higher performance level. This is depicted in Figure 3.

When both examinees are given the same rating, ( $F_{kk}$  times for a rating of  $k$ , and  $F_{ll}$  times for a rating of  $l$ ), their performance levels are indistinguishable. When the examinees are rated differently ( $F_{kl}$  and  $F_{lk}$

times), the examinee with the greater relative frequency of ratings in category k, the higher category, is perceived to have the higher ability. In comparing performance levels, we intend that the numeric result be independent of the number of replications. Thus, if the test were to be repeated again, and were of the same length, we would expect to get approximately the same numeric result. Moreover, if the two tests were then to be concatenated, we would again expect to obtain about the same result. The division of the two frequencies,  $F_{kl}$  and  $F_{lk}$ , is compatible with this expectation because we expect this ratio to be about the same when the test is repeated, and also when the two tests are concatenated. Consequently, the comparative levels of performance of examinees  $O_n$  and  $O_m$  can be identified by the stated ratio,  $F_{kl}/F_{lk}$ .

$$\frac{\text{Performance level of } O_n}{\text{Performance level of } O_m} = \frac{F_{kl}}{F_{lk}} \quad (7)$$

In the limit, the ratio of the empirically observed frequencies,  $F_{kl}/F_{lk}$ , becomes the ratio of the probabilities,  $P_{kl}/P_{lk}$ , where  $P_{kl}$  is the probability of examinee  $O_n$  being given a rating of k and examinee  $O_m$  a rating of l on one replication of item i, and  $P_{lk}$  is similarly defined. Thus we define the ratio  $P_{kl}/P_{lk}$  to be the ratio of the examinee's performances.

$$\frac{\text{Performance level of } O_n}{\text{Performance level of } O_m} = \frac{P_{kl}}{P_{lk}} \quad (8)$$

But, for objectivity, the ratings given examinees  $O_m$  and  $O_n$  must be independently awarded by the judge. Consequently,

$$P_{kl} = P_{nijk} * P_{mijl} \quad (9)$$

and

$$P_{lk} = P_{mijl} * P_{nijk} \quad (10)$$

where  $P_{nijk}$  is the probability of examinee  $O_n$  being given a rating of k on item  $A_i$  by judge  $J_j$ , and  $P_{mijl}$ ,  $P_{mijk}$ ,  $P_{mijl}$  are similarly defined. Then

$$\frac{\text{Performance level of } O_n}{\text{Performance level of } O_m} = \frac{P_{kl}}{P_{lk}} = \frac{F_{nijk}}{P_{nijl}} * \frac{P_{mijl}}{P_{mijk}} \quad (11)$$

However, also for objectivity, the relative performance of examinees  $O_n$  and  $O_m$  must be independent of which particular item is used to compare them. Thus, though performance levels are initially defined in terms of item  $A_i$ , the relative performance levels must have the same value when defined in terms of any conceptually equivalent item  $A_i'$ . That is

$$\frac{\text{Performance level of } O_n}{\text{Performance level of } O_m} = \frac{P_{nijk}}{P_{mijl}} * \frac{P_{n'jk}}{P_{m'jl}} = \frac{P_{n'jk}}{P_{m'jl}} * \frac{P_{nijk}}{P_{mijk}} \quad (12)$$



then

$$\frac{P_{nijk}}{P_{nijl}} = \frac{P_{mijk}}{P_{mijl}} * \frac{P_{ni'jk}}{P_{ni'jl}} * \frac{P_{mi'jl}}{P_{mi'jk}} \quad (13)$$

For objectivity, this ratio of the probabilities of examinee  $O_n$  being rated in categories  $k$  and  $l$  must be independent of whichever examinee  $O_m$  is used in the comparison. So, let us consider examinee  $O_0$  with performance level at the local origin of the ability scale. Similarly the ratio must also be independent of whichever item  $A_i'$  is used for the comparison. Thus it must also be hold for item  $A_0$  chosen to have difficulty at the local origin of the item scale.

$$\frac{P_{nijk}}{P_{nijl}} = \frac{P_{0ijk}}{P_{0ijl}} * \frac{P_{n0jk}}{P_{n0jl}} * \frac{P_{00jl}}{P_{00jk}} \quad (14)$$

If, instead of comparing performance levels by means of items  $A_i$  and  $A_i'$ , we compare performance levels by means of the ratings given by judges  $C_j$  and  $C_j'$  over numerous replications of item  $A_i$ , then again we expect the relative performance levels of the examinees to be maintained.

$$\frac{\text{Performance level of } O_n}{\text{Performance level of } O_m} = \frac{P_{nijk}}{P_{nijl}} * \frac{P_{mijl}}{P_{mijk}} = \frac{P_{nij'k}}{P_{nij'l}} * \frac{P_{mij'l}}{P_{mij'k}} \quad (15)$$

so that

$$\frac{P_{nijk}}{P_{nijl}} = \frac{P_{mijk}}{P_{mijl}} * \frac{P_{nij'k}}{P_{nij'l}} * \frac{P_{mij'l}}{P_{mij'k}} \quad (16)$$

Again this must be true if judge  $C_j'$  is chosen to be judge  $C_0$  with severity at the local origin of the severity scale, and examinee  $O_m$  is examinee  $O_0$ , and when item  $A_i$  is replaced by item  $A_0$ . Therefore

$$\frac{P_{n0jk}}{P_{n0jl}} = \frac{P_{00jk}}{P_{00jl}} * \frac{P_{n00k}}{P_{n00l}} * \frac{P_{000l}}{P_{000k}} \quad (17)$$

Furthermore, for objectivity, the relative severity levels of judges  $C_j$  and  $C_j'$  must be maintained whether the judging takes place over numerous replications of the administration of either item  $A_i$  or item  $A_i'$  to the same examinee  $O_n$ .

$$\frac{\text{Severity level of } C_j'}{\text{Severity level of } C_j} = \frac{P_{nijk}}{P_{nijl}} * \frac{P_{ni'jl}}{P_{ni'jk}} = \frac{P_{nij'k}}{P_{nij'l}} * \frac{P_{ni'j'l}}{P_{ni'j'k}} \quad (18)$$

then



$$\frac{P_{nijk}}{P_{nijl}} = \frac{P_{ni'jk}}{P_{ni'jl}} * \frac{P_{nij'k}}{P_{nij'l}} * \frac{P_{ni'j'l}}{P_{ni'j'k}} \quad (19)$$

Again this must be true if judge  $C_j'$  is judge  $C_0$  chosen at the origin of the severity scale, and examinee  $O_n$  is examinee  $O_0$ , and item  $A_i'$  is item  $A_0$ .

$$\frac{P_{0ijk}}{P_{0ijl}} = \frac{P_{00jk}}{P_{00jl}} * \frac{P_{0i0k}}{P_{0i0l}} * \frac{P_{000l}}{P_{000k}} \quad (20)$$

Substituting (20) and (17) in (14), and simplifying

$$\frac{P_{nijk}}{P_{nijl}} = \frac{P_{n00k}}{P_{n00l}} * \frac{P_{0i0k}}{P_{0i0l}} * \frac{P_{00jk}}{P_{00jl}} * \frac{(P_{0000l})^2}{(P_{0000k})} \quad (21)$$

which gives a general form in which each term is an expression of the relationship between a component of a facet and the local origin of a subscale, in the context of a particular pair of categories.

#### IV. The three-faceted dichotomous model.

Considering (21) as a dichotomous model where  $k=1$  ("right") and  $l=0$  ("wrong"), then this equation expresses the ratio of the probabilities of the possible outcomes as a product of terms which relate each component of each facet with the local origin of its subscale. These terms are independent of whichever other examinees, items and judges are included in the test situation. To consider these terms in an additive way, we can take logarithms and assign a numerical direction to each term in accordance with conventional interpretation. Let

$B_n = \log(P_{n001}/P_{n000})$ , which is defined to be the ability of an examinee  
 $D_i = \log(P_{0i00}/P_{0i01})$ , which is defined to be the difficulty of an item  
 $C_j = \log(P_{00j0}/P_{00j1})$ , which is defined to be the severity of a judge

We also define the relationship of the subscales, such that the probability of "original" examinee  $O_0$  being rated a "1" by judge  $C_0$  on item  $A_0$  is 0.5, so that the last term of equation (21) becomes 0.

Reparameterizing  $P_{nij} = P_{nij1}$ , so that  $1 - P_{nij} = P_{nij0}$ , gives equation (22), the three-faceted Rasch model for the dichotomous case.

$$\log(P_{nij}/(1-P_{nij})) = B_n - D_i - C_j \quad (22)$$

#### V. The three-faceted rating scale model.

When we consider examinees  $O_n$  and  $O_m$  in the more general circumstances of a rating scale, we do not wish the comparison of their abilities to depend on which particular pair of categories of the rating scale are used for the

comparison. So we return to equation (11) which stated:

$$\frac{\text{Performance level of } O_n}{\text{Performance level of } O_m} = \frac{P_{kl}}{P_{lk}} = \frac{P_{nijk}}{P_{nijl}} * \frac{P_{mijl}}{P_{mijk}} \quad (11)$$

We wish to generalize this equation to any pair of categories, but the rating scale categories are not independent but structured. In order to determine the structure in an objective manner, we require that performance levels are invariant when they are compared using any pair of adjacent categories in ascending order. This is the only possible objective structuring since invariance, which is not over adjacent categories, but over some pairing of non-adjacent categories results in a contradiction or indeterminacy in the rating scale structure. Thus, if a rating scale has 3 categories and performance levels are to be invariant only when the top and bottom categories are used for the comparison in (11), then performance levels based on the middle category are indeterminate, and are not objective.

Invariance in relative performance when categories are chosen such that  $k$  is one greater than  $l$ , and also  $k'$  is chosen one greater than  $l'$ , yields

$$\frac{\text{Performance level of } O_n}{\text{Performance level of } O_m} = \frac{P_{nijk}}{P_{nijl}} * \frac{P_{mijl}}{P_{mijk}} = \frac{P_{nijk'}}{P_{nijl'}} * \frac{P_{mijl'}}{P_{mijk'}} \quad (23)$$

so that

$$\frac{P_{nijk}}{P_{nijl}} = \frac{P_{nijk'}}{P_{nijl'}} * \frac{P_{mijk}}{P_{mijl}} * \frac{P_{mijl'}}{P_{mijk'}} \quad (24)$$

Since we want this result to be generalizable, we must be able to substitute examinee  $O0$ , item  $A0$ , and judge  $C0$ ,

$$\frac{P_{n00k}}{P_{n00l}} = \frac{P_{n00k'}}{P_{n00l'}} * \frac{P_{000k}}{P_{000l}} * \frac{P_{000l'}}{P_{000k'}} \quad (25)$$

Reordering the terms,

$$\frac{P_{n00k}}{P_{n00l}} = \left( \frac{P_{n00k'}}{P_{n00l'}} * \frac{P_{000l'}}{P_{000k'}} \right) * \frac{P_{000k}}{P_{000l}} \quad (26)$$

the two terms in parentheses are invariant over changes in choice of pairs of categories and so are independent of the local structure of the rating scale, but they are not independent of the choice of object, so we can accordingly rewrite them as  $P_{n00}$ , so that

$$\frac{P_{n00k}}{P_{n00l}} = P_{n00} * \frac{P_{000k}}{P_{000l}} \quad (27)$$

Similar equations hold for  $P0i0k/P0i0l$  and  $P00jk/P00jl$ , so that, substituting into (21),

$$\frac{Pnijk}{Pnijl} = Pn00 * P0i0 * P00j * \frac{P000k}{P000l} \quad (28)$$

We have an equation in which the ratio of the probabilities of particular outcomes is the product of terms which depend only on a single component and the local origin of its subscale, combined with a term dependent on the pair of categories used for the comparison. Let

$B_n = \log(Pn00)$ , which is defined to be the examinee ability,  
 $D_i = -\log(P0i0)$ , which is defined to be the item difficulty,  
 $C_j = -\log(P00j)$ , which is defined to be the judge severity  
 $F_k = -\log(P000k/P000l)$ , which is defined to be the difficulty of the step from category  $k-1$  ( $=l$ ) to category  $k$  of the rating scale.

Then equation (28), the three-faceted Rasch rating scale model becomes

$$\log(Pnijk/Pnijk-1) = B_n - D_i - C_j - F_k \quad (29)$$

In (29), the parameters relating to the particular examinees, items, and judges interacting to make each rating have been separated, and so (29) is objective in that the parameters of the particular examinees, judges and items enter independently into the rating process. Nevertheless, the parameter estimate of  $B_n$ , say, could be inflated by an arbitrary amount so long as the other parameter estimates were deflated accordingly. Thus the actual placement of the examinee, item, judge and step subscales within the common frame of reference is arbitrary. By convention, local origins for each subscale are chosen such that the mean calibrations of the items, of the judges, and of the rating scale steps are each zero. The local origin of the examinee abilities' subscale is then defined uniquely by the model.

## VI. Conclusion

The manner in which the dichotomous two-faceted Rasch model can be derived from objectivity has been demonstrated to extend to an example of the many-faceted model which includes a rating scale. Any other particular form of the many-faceted model may also be derived in a similar way.

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